

**COMPARING THE PERFORMANCES OF NEURAL NETWORK AND ROUGH SET  
THEORY TO REFLECT THE IMPROVEMENT OF PROGNOSTIC IN MEDICAL  
DATA**

A thesis submitted to College of Arts and Sciences in partial fulfillment of the requirement for  
the degree Master of Science (Intelligent System) Universiti Utara Malaysia

By

Nur Aniza Bt Alang Ismail

December, 2009

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## ABSTRAK

Melalui penyelidikan singkat yang telah saya jalankan, saya telah menyelidik dua daripada teknik yang telah diperkenalkan dalam Kepintaran Buatan; iaitu Rangkaian Neural (Neural Network) dan juga Teori Set Kasar (Rough Set Theory). Kedua-dua teknik ini adalah dua teknik yang terbaik digunakan dalam penganalisaan data. Kepintaran Buatan adalah merupakan satu teknik yang masih deperingkat awal dan ianya baru diperkenalkan. Ianya masih lagi diperingkat pembangunan dan kegunaannya adalah menghasilkan sistem pintar yang dapat membantu manusia dalam kehidupan seharian bagi menyokong proses dalam membuat satu-satu keputusan.

Di Malaysia, Kepintaran Buatan adalah satu bidang yg masih lagi baru diperkenalkan. Satu kumpulan penyelidik dari Universiti Sains Malaysia telah menjalankan kajian tentang Kepintaran Buatan ini dalam bidang perubatan. Mereka juga bersetuju dengan kenyataan yang diberikan oleh para penyelidik Kepintaran Buatan seluruh negara bahawa Kepintaran Buatan sangat membantu dalam menggantikan kepintaran manusia. Dengan adanya elemen Kepintaran Buatan, ia membantu menyelesaikan pelbagai tugas manusia terutamanya dalam bidang perubatan dan disamping itu juga dapat mempercepatkan proses kerja seharian.

Dalam penyelidikan saya ini, saya telah memilih tiga set data perubatan iaitu; Ramalan Kanser Payudara dari Wisconsin, Penyakit Parkinson dan ramalan penyakit Hepatitis. Data-data perubatan ini telah dipilih kerana data-data yang berkaitan dengan perubatan seringkali digunakan oleh penyelidik-penyelidik Kepintaran Buatan dalam menjalankan kajian mereka. Selain itu, keputusan ramalan dan juga data-data yang digunakan dalam kajian ini mudah difahami. Selain itu, metodologi yang digunakan untuk kajian ini turut dibincangkan dan saya juga telah membuat kesimpulan dan juga kajian yang bakal dijalankan sebagai satu kesinambungan daripada kajian yang telah dijalankan ini.

## **ABSTRACT**

In this research, I investigate and compared two of Artificial Intelligence (AI) techniques which are; Neural network and Rough set will be the best technique to be use in analyzing data. Recently, AI is one of the techniques which still in development process that produced few of intelligent systems that helped human to support their daily life such as decision making. In Malaysia, it is newly introduced by a group of researchers from University Science Malaysia. They agreed with others world-wide researchers that AI is very helpful to replaced human intelligence and do many works that can be done by human especially in medical area.

In this research, I have chosen three sets of medical data; Wisoncin Prognostic Breast cancer, Parkinson's diseases and Hepatitis Prognostic. The reason why the medical data is selected for this research because of the popularity among the researchers that done their research in AI by using medical data and the prediction or target attributes is clearly understandable. The results and findings also discussed in this paper. How the experiment has been done; the steps involved also discussed in this paper. I also conclude this paper with conclusion and future work.

## **ACKNOWLEDGEMENT**

Alhamdulillah, it is with Allah S.W.T will that I get finish this Final Project in order to complete my Master's degree. I am very thankful to Dr. Fauziah bt Ahmad whom has been supervised me throughout this semester to complete this Master's Thesis. Also special thanks to Miss Aniza bt Mohamed Din whom helped me a lot in giving guidance and information in performing this thesis. Not forgetting my family for their support and understanding.

This paper is focusing in Artificial Intelligence techniques; Neural Network and Rough Set technique in order to get the best technique to be use in analyzing data. The software that has been used in the experiment is Neural Connection and ROSETTA.

*“To my beloved family, thanks for your support and sacrifice. To all my friends, nice knowing you all and thanks for the understanding and encouragement.”*

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**LIST OF ABBREVIATIONS**

AI	Artificial Intelligence
KDD	Knowledge Data and Discovery
NN	Neural Network
RS	Rough Set
OLAP	Online Analytical processing
HBV	Hepatitis virus B
WHO	World Health Organization
HBsAg	Hepatitis B surface Antigen
PNN	Probabilistic Neural Network
FNA	Fine needle aspirate
DM	Data Mining
MLP	Multilayer Perceptron
RMS	Root Mean Squared
GA	Genetic Algorithm



## **CHAPTER ONE: INTRODUCTION**

Artificial Intelligence is one of approach that can train computers to think like human, where it can learn through experience, recognize patterns from large amount of data and also decision making process based from human knowledge and reasoning skills. According from an AI text book titled AI: Structures and Strategies for Complex Problem Solving, an AI can be defined as the branch of computer science that is concerned with the automation of intelligent behavior (Luger, 2005). It is combination of science and engineering field in order to make an intelligent machines, especially intelligent computer programs. There are three (3) perspectives in AI; 1) AI can be as a replacement, 2) as an assistant and 3) it also can be used to extend human capabilities (McCarthy J., 2007).

Nowadays, computers technology and data bases helps human in collecting and storing huge amount of data. The large size of most data bases makes it impossible for human to interpret data. Therefore, computers are needed for extracting new, useful knowledge. Lately, other science methods like machine learning, artificial intelligence and logics have made progress and achievements in this field. Today, as we can see the usage of Data Mining and Knowledge Discovery gives more advantages to Statisticians in order to reduce the information stored, to reduce costs, increase sales and revenues, also reduce accidents and failure within data (Dingsoyr T., 1997). There too many definitions about Data Mining and Knowledge Discovery

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